

Linear Regression

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Outline

- Simple Linear Regression (SLR)
 - Model
 - Interpretation
 - Assumptions
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Simple Linear Regression

For observation i , let

- Y_i be the response variable and
- X_i be the explanatory variable.

The simple linear regression model (SLR) assumes

$$Y_i \stackrel{ind}{\sim} N(\beta_0 + \beta_1 X_i, \sigma^2)$$

or, equivalently,

$$Y_i = \beta_0 + \beta_1 X_i + \epsilon_i, \quad \epsilon_i \stackrel{ind}{\sim} N(0, \sigma^2).$$

Interpretation

Recall

$$E[Y_i] = \beta_0 + \beta_1 X_i$$

Thus,

- β_0 is the expected response when $X_i = 0$
- β_1 is the expected increase in the response when X_i is increased by 1.

Assumptions

Recall

$$E[Y_i] = \beta_0 + \beta_1 X_i, \quad \epsilon_i \stackrel{ind}{\sim} N(0, \sigma^2)$$

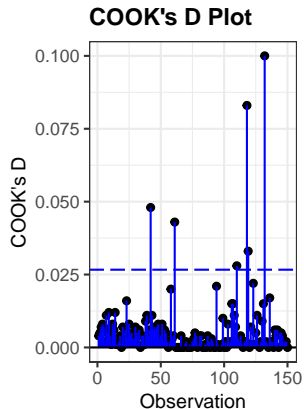
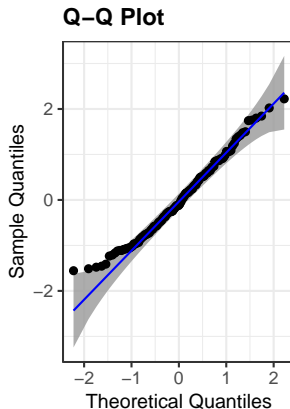
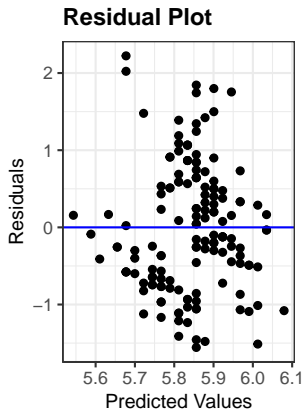
Thus, the model assumptions are

- The errors are normally distributed.
- The errors have constant variance.
- The errors are independent.
- The relationship between the expected response and the explanatory variable is a straight line.

Diagnostics

To evaluate these model assumptions we utilize diagnostic plots:

```
m <- lm(Sepal.Length ~ Sepal.Width, data = iris)
ggResidpanel::resid_panel(m, plots = c("resid", "qq", "cookd"), qqbands = TRUE, nrow = 1)
```



Triathlon Data

from <https://modules.scorenetwork.org/triathlons/ironman-lakeplacid-mlr/>

```
d <- read_csv("ironman_lake_placid_female_2022_canadian.csv")
```

```
head(d)
```

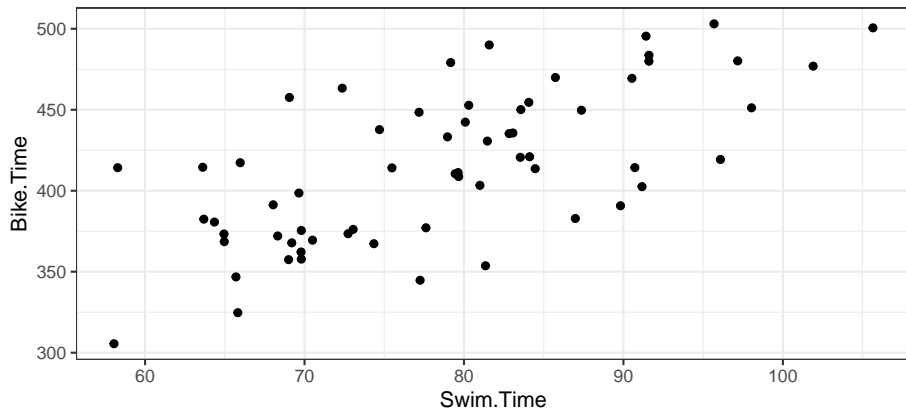
```
# A tibble: 6 x 17
```

	Bib	Name	Country	Gender	Division	Division.Rank	Overall.Time	Overall.Rank	Swim.Time	Swim.Rank	Bike
	<dbl>	<chr>	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	2	Melanie~	Canada	Female	FPR0	5	575.	21	58.0	57	
2	9	Pamela-~	Canada	Female	FPR0	10	610.	51	65.8	253	
3	1000	Carley ~	Canada	Female	F35-39	4	660.	126	65.7	249	
4	1935	Seanna ~	Canada	Female	F45-49	3	665.	131	74.4	727	
5	511	Marie-C~	Canada	Female	F45-49	4	679.	161	77.2	899	
6	1240	Julie H~	Canada	Female	F40-44	6	693.	202	77.6	921	

i 5 more variables: Run.Time <dbl>, Run.Rank <dbl>, Finish.Status <chr>, Location <chr>, Year <dbl>

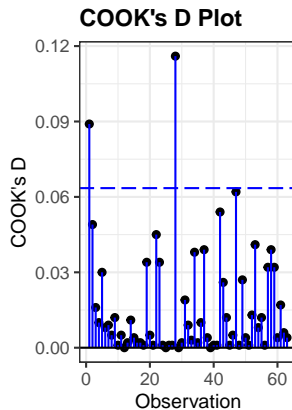
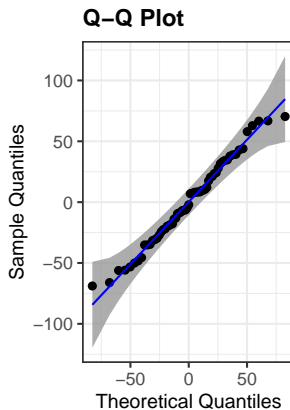
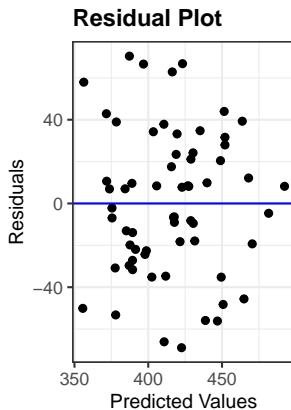
Bike Time v Swim Time

```
ggplot(d |> filter(Swim.Time < 500), aes(x = Swim.Time, y = Bike.Time)) + geom_point()
```



Bike Time v Swim Time - Model Diagnostics

```
m <- lm(Bike.Time ~ Swim.Time, data = d |> filter(Swim.Time < 500))  
ggResidpanel::resid_panel(m, plots = c("resid", "qq", "cookd"), qqbands = TRUE, nrow = 1)
```



Bike Time v Swim Time - Model Results

```
summary(m)
```

Call:

```
lm(formula = Bike.Time ~ Swim.Time, data = filter(d, Swim.Time <
  500))
```

Residuals:

Min	1Q	Median	3Q	Max
-68.901	-23.468	-2.169	23.808	70.369

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	188.8604	32.0893	5.885	1.82e-07 ***
Swim.Time	2.8729	0.4035	7.120	1.44e-09 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 34.83 on 61 degrees of freedom

Multiple R-squared: 0.4538, Adjusted R-squared: 0.4449

F-statistic: 50.69 on 1 and 61 DF, p-value: 1.443e-09

Bike Time v Swim Time - Written Results

```
cbind(coef(m), confint(m))
```

		2.5 %	97.5 %
(Intercept)	188.860386	124.693942	253.026829
Swim.Time	2.872855	2.065987	3.679724

```
summary(m)$r.squared
```

```
[1] 0.4538433
```

When swim time is 0, the expected Bike Time is 189 mins with a 95% interval of (125, 253). For additional minute of swim time, the bike time is expected to increase 2.9 mins (2.1, 3.7). The model explains 45% of the variability in bike time.

Bike Time v Swim Time - Plot

```
ggplot(d |> filter(Swim.Time < 500), aes(x = Swim.Time, y = Bike.Time)) +  
  geom_point() + geom_smooth(method = "lm")
```

